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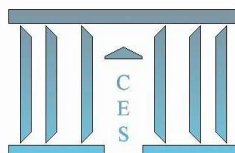
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**Optimal Rationing within a Heterogeneous Population**

Philippe CHONÉ, Stéphane GAUTHIER

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# Optimal Rationing within a Heterogeneous Population

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## Abstract

A government agency delegates to a provider (hospital, medical gatekeeper, school, social worker) the decision to supply a service or treatment to individual recipients. The agency does not perfectly know the distribution of individual treatment costs in the population. The single-crossing property is not satisfied when the uncertainty pertains to the dispersion of the distribution. We find that the provision of service should then be distorted upwards relative to efficiency when the (first-best) efficient number of recipients is sufficiently high.

**JEL classification numbers:** I18, D82, D61

**Keywords:** Rationing, screening, universal coverage, upward distortion, Spence-Mirrlees condition

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# 1 Introduction

We consider a government agency in charge of supplying a service or treatment to a population of potential recipients. We assume that the cost and the benefit of the treatment vary across individuals. Efficiency requires that the treatment is provided to an individual if and only if social benefit exceeds social cost for that individual.

The individual heterogeneity in cost and benefit may in part be explained by recipients' characteristics that are observed by the agency. To some extent, the agency can thus use "rationing by denial", whereby the treatment is denied to individuals with unfavorable cost-benefit ratio. In many instances, however, there remains substantial unobserved heterogeneity conditional on observable variables. The agency has then to leave the selection of recipients to the discretion of the service provider, i.e. rely on "rationing by selection", see [3]. In practice, the supply decision is indeed delegated to the provider for various services, e.g. medical or social care, after-school or training programs.

Under rationing by selection, the agency first determines the total number of recipients. This step requires knowing the distribution of unobserved heterogeneity among potential recipients. Perfect knowledge of the distribution is commonly assumed in the literature. For instance, the assumption is made in [5] and [4], with the former article considering many treatment varieties and the latter investigating provider altruism –two issues not addressed here.

The present note places the emphasis on the uncertainty about the distribution of unobserved heterogeneity. The composition of the population of recipients addressed by a given provider depends on local conditions, creating significant variation in the underlying distributions across providers. The effect of differences in composition on the cost and benefit distributions is difficult to estimate empirically (see [6] for a recent survey).

Accordingly, we relax the assumption that the agency perfectly knows the cost and benefit distribution in the population of potential recipients. We find that the uncertainty about the distribution of heterogeneity causes the number of recipients to be distorted relative to efficiency. The direction of the distortion depends on whether the uncertainty pertains to the mean or to the dispersion of individual treatment costs. In the former case, the usual Spence-Mirrlees condition is satisfied and the distortion is necessarily downwards: The number of treated recipients is lower than recommended

by first-best efficiency. In the latter case, the Spence-Mirrlees condition no longer holds. We find that the first-best optimum governs the sign of distortion in the second-best program: The distortion is upwards when the first-best number of treated recipients is sufficiently high. Uncertainty about the cost dispersion then pushes towards universal coverage policies.

## 2 Model

A population of recipients, whose size is normalized to one, is eligible for a treatment supplied by a single provider. Recipients are indexed by two nonnegative real numbers  $b$  and  $c$ , with distribution function  $\Phi(b, c)$ . The treatment of a type  $(b, c)$  recipient yields benefit  $b$  to the recipient and costs  $c$  to the provider. The corresponding net social benefit is  $b - (1 + \lambda)c$ , where  $\lambda$  is the (exogenously given) marginal cost of public funds.

**Assumption 1.** *The (expected) net social benefit of treatment for a given cost level  $c$ ,  $\mathbb{E}(b \mid c) - (1 + \lambda)c$ , is a non-increasing function of cost  $c$  with a unique zero, denoted by  $c^{**}$ .*

Assumption 1 holds true in particular when the expected benefit decreases with cost, a case often considered in the literature, e.g., in [2] and [4]. [5] provides health related examples where this assumption is relevant. Under this assumption, the first-best requires to treat recipients with cost  $c \leq c^{**}$ . Denoting by  $F$  the marginal distribution of individual treatment costs in the population of recipients, the first-best optimal number of treated recipients is  $n^{**} = F(c^{**})$ .

In this paper, we assume that the agency relies on rationing by selection: the agency observes the number  $n$  of recipients but not their individual characteristics. The treatment decision is delegated to a single provider who observes the individual characteristics of recipients. The agency offers a take or leave contract specifying the number of recipients that must be treated by the provider and a compensating transfer  $T$ . The utility of the provider when she accepts the contract is  $U(n, T) = T - C(n)$ , where  $C(n)$  represents the aggregate cost of treating  $n$  recipients.

Given  $(n, T)$ , utility maximization requires that the least costly recipients be treated in priority. The provider's cost of treating  $n$  recipients,  $0 \leq n \leq 1$ ,

is therefore given by

$$C(n) = \int_0^{F^{-1}(n)} c \, dF(c).$$

The marginal cost is  $C'(n) = F^{-1}(n)$ , i.e., the cost of the marginal treated recipient is  $F^{-1}(n)$ , the  $n$ th-percentile of the distribution  $F$ . It follows that the cost function  $C(n)$  is convex in  $n$ .

The net social benefit of treating  $n$  recipients is given by  $S(n) = B(n) - (1 + \lambda)C(n)$ , where

$$B(n) = \int_0^{F^{-1}(n)} \mathbb{E}(b \mid c) \, dF(c).$$

represents the (expected) the gross social benefit. Under Assumption 1, the net social benefit function  $S(n)$  is concave in  $n$ , reaching its maximum at  $n^{**} = F(c^{**})$ . When the agency knows the distribution of individual recipients characteristics  $\Phi$ , it can choose the number of recipients  $n$  and the transfer  $T$  that maximize

$$B(n) + U(n, T) - (1 + \lambda)T = S(n) - \lambda U(n, T)$$

subject to the provider's participation constraint  $U(n, T) \geq 0$ . The solution to this maximization problem is the first-best optimum,  $n = n^{**}$  and  $U = U^{**} = 0$ .

Under Assumption 1, the provider treats in priority the recipients with highest expected social values: The social net benefit and the provider's private objective are 'aligned.' Hence, the agency can achieve the first-best optimum if it knows the distribution of individual recipients characteristics: From the distribution function  $\Phi$ , the agency infers the optimal number  $n^{**}$  and the corresponding cost,  $C(n^{**})$ . It is then sufficient to ask the provider to treat  $n^{**}$  recipients and reimburse  $C(n^{**})$ .

### 3 Unknown cost distribution

We now relax the assumption that the agency knows the distribution of individual characteristics of recipients. We suppose that the agency only knows that the distribution function is  $\Phi_i$  ( $i = H, L$ ), associated with marginal cost distribution  $F_i$ , with probability  $\pi_i$  ( $\pi_L + \pi_H = 1$ ). The true distribution function is private information to the provider. We refer to  $i$  as the provider

type. Hence, the agency is faced with two types of provider, themselves confronted with different populations of recipients and marginal distributions of individual costs. Provider  $i$  has cost function  $C_i(n)$  with  $C'_i(n) = F_i^{-1}(n)$ , and the net social benefit of having  $n$  recipients treated by provider  $i$  is

$$S_i(n) = \int_0^{F_i^{-1}(n)} [\mathbb{E}_i(b | c) - (1 + \lambda)c] dF_i(c),$$

where the conditional expectation  $\mathbb{E}_i$  is taken under provider  $i$ 's distribution  $\Phi_i$ . Assumption 1 is supposed to hold for the two provider types: The first-best cost threshold for provider  $i$  is  $c_i^{**}$  such that  $\mathbb{E}_i(b | c_i^{**}) = (1 + \lambda)c_i^{**}$ , where  $\mathbb{E}_i(b | c_i^{**})$  represents the mean treatment benefit for provider  $i$  in the population of recipients whose treatment cost is  $c_i^{**}$ . The corresponding first-best number of recipients is  $n_i^{**} = F_i(c_i^{**})$ .

Appealing to the revelation principle, the agency offers a menu  $(n_i, T_i)$ ,  $i = H, L$ , maximizing

$$\sum_i \pi_i [S_i(n_i) - \lambda U_i(n_i, T_i)]$$

subject to the provider's participation constraints  $U_i(n_i, T_i) = T_i - C_i(n_i) \geq 0$  and the incentive constraints  $U_i(n_i, T_i) \geq U_i(n_j, T_j)$  for all  $i, j = H, L$ .

We examine how the solution to this problem departs from the first-best optimum when the marginal distribution  $F_L$  stochastically dominates the distribution  $F_H$ .

### 3.1 Local distortions

Suppose first that  $F_L$  first-order stochastically dominates  $F_H$ :  $F_L(c) \leq F_H(c)$  for all  $c$ , with strict inequality on a non-degenerated interval. In this case, for all  $n > 0$ , provider  $H$ 's marginal cost stands below provider  $L$ 's:

$$C'_H(n) = F_H^{-1}(n) < C'_L(n) = F_L^{-1}(n),$$

and hence  $C_H(n) < C_L(n)$ , since  $C_H(0) = C_L(0)$ . The first-best menu  $(n_i^{**}, C_i(n_i^{**}))$  is not incentive compatible: Provider  $H$  chooses  $(n_L^{**}, C_L(n_L^{**}))$  and earns  $C_L(n_L^{**}) - C_H(n_L^{**}) > 0$ . In order to reduce the informational rent  $C_L(n_L) - C_H(n_L)$  of the type  $H$  provider, the agency distorts the number

of recipients treated by type  $L$  with respect to the first-best  $n_L^{**}$ . The usual single-crossing condition is satisfied:

$$\left. \frac{\partial T}{\partial n} \right|_{U_H} = C'_H(n) < C'_L(n) = \left. \frac{\partial T}{\partial n} \right|_{U_L} \quad \text{for all } n.$$

Therefore a lower rent requires a lower number of recipients treated by provider  $L$ . This is the standard pattern in adverse selection problems.

**Proposition 1.** *Suppose that  $F_L$  first-order stochastically dominates  $F_H$ . At the second-best optimum, provider  $H$  (provider  $L$ ) treats the optimal number of (too few) recipients and earns a positive (zero) profit.*

*Proof.* Since  $C_L(n) > C_H(n)$ , the type  $H$  incentive constraint is binding,  $U_H^* = C_L(n_L^*) - C_H(n_L^*)$ , and  $U_L^* = 0$ . The optimal number  $n_L^*$  satisfies the first-order condition

$$\pi_L S'_L(n_L^*) - \lambda \pi_H [C'_L(n_L^*) - C'_H(n_L^*)] = 0. \quad (1)$$

Thus  $S'_L(n_L^*) > 0 = S'(n_L^{**})$ , and  $S(n)$  being single-peaked,  $n_L^* < n_L^{**}$ . Type  $H$  treats  $n_H^* = n_H^{**}$  recipients and earns  $U_H^* = C_L(n_L^*) - C_H(n_L^*) > 0$ .  $\square$

The main result of the paper concerns the case where  $F_H$  is a mean-preserving spread of  $F_L$ : Both distributions  $F_H$  and  $F_L$  have the same mean,

$$C_L(1) = \int_0^\infty c \, dF_H(c) = \int_0^\infty c \, dF_L(c) = C_H(1),$$

and  $F_L$  second-order stochastically dominates  $F_H$ ,

$$\int_0^c F_L(c) \, dc \leq \int_0^c F_H(c) \, dc \quad \text{for all } c.$$

**Assumption 2.** *The two distribution functions cross only once, at some individual treatment cost denoted  $\hat{c}$ .*

The single-crossing property does not hold, but Assumption 2 restricts the pattern of violation of this assumption, delimitating two different regions:

$$\left. \frac{\partial T}{\partial n} \right|_{U_H} = C'_H(n) < C'_L(n) = \left. \frac{\partial T}{\partial n} \right|_{U_L} \iff n < \hat{n}, \quad (2)$$



with  $\hat{n} = F_L(\hat{c}) = F_H(\hat{c})$ . In the spirit of [1], we use this partition to handle the adverse selection problem.<sup>1</sup>

**Proposition 2.** *Suppose that  $F_H$  is a mean-preserving spread of  $F_L$  and that Assumption 2 holds. Then, provider's  $H$  incentive constraint is the only one to be binding at the optimum. This provider treats  $n_H^{**}$  recipients.*

*The agency has a local incentive to distort the number of recipients treated by provider  $L$  upwards from  $n_L^{**}$  if and only if  $n_L^{**} > \hat{n}$ .*

*Proof.* By Assumption 2,  $F_H(c) > F_L(c)$  for  $c < \hat{c}$  and  $F_H(c) < F_L(c)$  for  $c > \hat{c}$ . Using the expression of the marginal costs, we find that  $C'_H(n) < C'_L(n)$  for  $n < \hat{n}$  and  $C'_H(n) > C'_L(n)$  for  $n > \hat{n}$ . Since  $C_H(0) = C_L(0)$  and  $C_H(1) = C_L(1)$ , we have

$$C_H(n) \leq C_L(n)$$

for all  $n$ , with equality for  $n = 0$  and  $n = 1$ . The cost difference  $C_L(n) - C_H(n)$  first increases, then decreases as  $n$  rises, achieving its maximum at  $\hat{n} = F_H(\hat{c}) = F_L(\hat{c})$ . The efficient provider, provider  $H$ , is the one faced with more dispersed individual treatment costs.

It follows that at the second-best optimum,  $H$ 's incentive constraint must be binding,  $U_H^* = C_L(n_L^*) - C_H(n_L^*)$ , and  $U_L^* = 0$ . The number of recipients treated by provider  $H$  is undistorted. The optimal number  $n_L^*$  of treated by the  $L$  type maximizes

$$K(n) \equiv \pi_L S_L(n) + \lambda \pi_H [C_H(n) - C_L(n)]. \quad (3)$$

The first derivative of this function at  $n_L^{**}$  is  $\lambda \pi_H [C'_H(n_L^{**}) - C'_L(n_L^{**})]$ , since  $S_L(n_L^{**}) = 0$ . By (2), it is positive if and only if  $n_L^{**} > \hat{n}$ .  $\square$

Figures 1 and 2 describe how to derive the aggregate cost function from the marginal distribution of individual cost. By Assumption 2 the marginal distributions intersect at  $\hat{c}$ . From the second order stochastic dominance property we also know that  $F_H(c)$  is above (below)  $F_L(c)$  for  $c$  less (greater) than  $\hat{c}$ . Therefore, the slope of the cost function for provider  $H$  is less (greater) than the slope of the cost function for provider  $L$  when  $n$  is less (greater) than  $\hat{n}$ . This change in the way the slopes are ordered implies a failure of the usual single-crossing condition. However, given that both providers

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<sup>1</sup>In our setup, however,  $C'_H(n) - C'_L(n)$  is non-monotonic in  $n$ , hence [1]'s Assumption A2 does not hold.

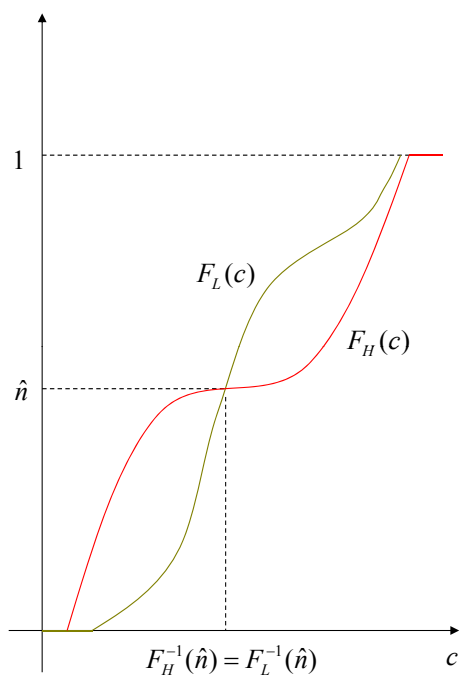


Figure 1: Individual cost cdf

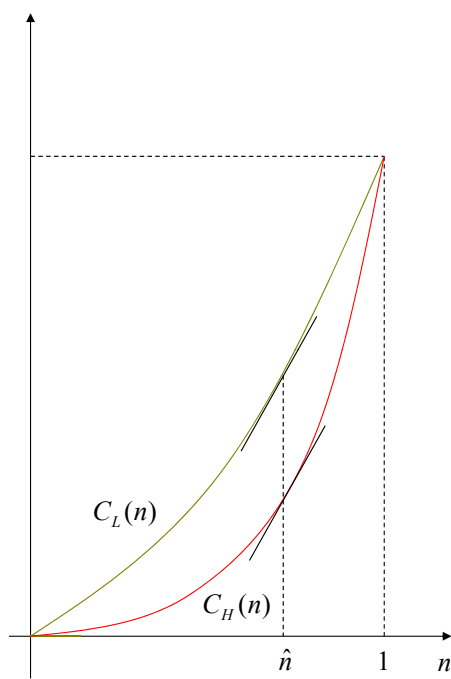


Figure 2: Aggregate cost function



has a local incentive to reduce the number of treated by the  $L$  type when  $n_L^{**} < \hat{n}$ , as is the case in Figure 3. On the contrary,  $n_L^{**} > \hat{n}$  in Figure 4, and the agency then has a local incentive to increase the number of treated above  $n_L^{**}$ .

Thus, unlike the main strand of the literature, the sign of the local distortion from the first-best optimum in the second-best setup are driven by the number of treated in the first-best problem: A high number of treated in the first-best yields an upward distortion.

### 3.2 A global result

Previous results describe the pattern of local incentives to distort the number of treated recipients from the first-best. However it could be that the global optimum involves to treat  $n_L^* < n_L^{**}$  recipients in the case where  $n_L^{**}$  is above  $\hat{n}$ . To go beyond the local result of Proposition 2, we introduce

**Assumption 3.** *The distributions  $F_L$  and  $F_H$  are symmetric around  $\hat{c}$ , i.e.  $F_i(c) + F_i(2\hat{c} - c) = 1$  for all  $c$  and  $i = H, L$ .*

Under Assumptions 2 and 3, the distributions are equal to one half when they cross:  $\hat{n} = F_H(\hat{c}) = F_L(\hat{c}) = 1/2$ .

**Proposition 3.** *Suppose that Assumptions 2 and 3 hold,  $F_H$  is a mean-preserving spread of  $F_L$ , and  $n_L^{**}$  is larger than  $1/2$ . Then the number of recipients treated by provider  $L$  is distorted upwards if  $S_L(1) > S_L(1 - n_L^{**})$ .*

*Proof.* By Assumption 3, the rent  $U_H(n_L) = C_L(n_L) - C_H(n_L)$  is symmetric around its global maximum  $\hat{n} = 1/2$ , i.e.  $U_H(n) = U_H(1 - n)$  for all  $n$ . Recall that the second-best number  $n_L^*$  of recipients treated by provider  $L$  maximizes the function  $K(n)$  given by (3). We want to show that  $n_L^* > n_L^{**}$  when  $n_L^{**} > 1/2$ . The proof proceeds in three steps:

1. Let  $\tilde{n}_L = 1 - n_L^{**} < 1/2$ . Since  $U_H(n) = U_H(1 - n)$  and  $S_L$  increases below  $n_L^{**}$ , the maximum of  $K$  on  $[\tilde{n}_L, n_L^{**}]$  is achieved above  $1/2$ .
2. Since  $S_L$  is increasing and  $U_H$  is decreasing on the interval  $[1/2, n_L^{**}]$ ,  $K$  is increasing on this interval, implying that the maximum of  $K$  on  $[\tilde{n}_L, n_L^{**}]$  is achieved at  $n_L^{**}$ .

3. By symmetry of  $U_H$  and monotonicity of  $S_L$  on the intervals  $[0, n_L^{**}]$  and  $[n_L^{**}, 1]$ , we have

$$K(n) - K(1 - n) = S_L(n) - S_L(1 - n) \geq S_L(1) - S_L(1 - n_L^{**}) > 0$$

for all  $n \geq n_L^{**}$ . It follows that the maximum of  $K(n)$  on  $[0, 1]$  is achieved at the right of  $n_L^{**}$ .

By Proposition 2, the maximum is achieved strictly above  $n_L^{**}$ . □

This global result is closely reminiscent of Proposition 2: The number of recipients is distorted upwards from the first-best optimum if the agency prefers the inefficient provider (the one faced with less dispersed individual treatment costs) to treat a sufficiently high number of recipients.

The new global condition in Proposition 3 depends on whether treating all the population is socially preferred to treating only a fraction  $1 - n_L^{**}$ . It should be satisfied in practice: if the first-best optimum recommends to treat  $n_L^{**} = 80\%$  of the population, the agency should prefer providing universal coverage to treating only  $1 - n_L^{**} = 20\%$  of the population.<sup>2</sup>

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<sup>2</sup>This condition follows from Assumption 3. Global results involving an upwards distortion obtain for asymmetric distributions when, for high values of  $n$  ( $n > \hat{n}$ ), the first-best surplus for the inefficient provider is sufficiently high, and the difference between the aggregate cost functions of both providers decreases sufficiently with  $n$ .

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